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METHOD OF INSPECTING CLEANLINESS OF TOP AND DEVICE USED THEREFOR.

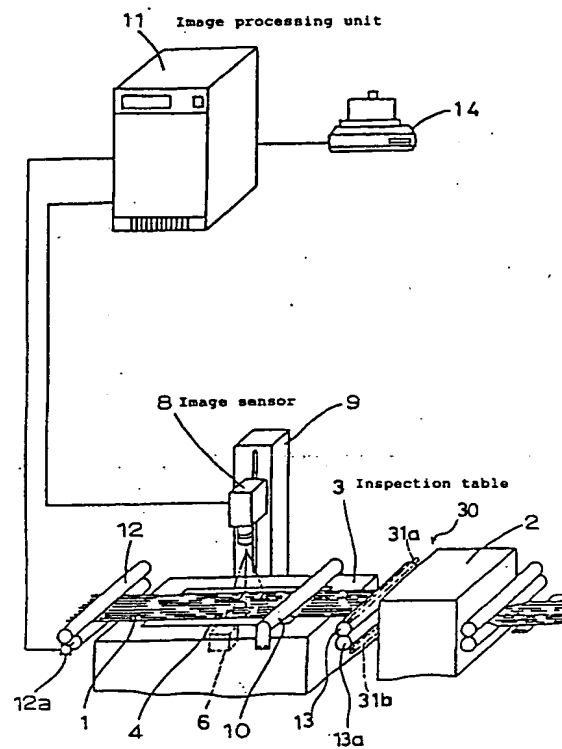
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A method of inspecting cleanliness of a top in which pill defects and plant defects are classified from image signals obtained through imaging the top with an image sensor by adopting suitably the methods of discrimination, such as roundness discrimination, slenderness discrimination, gradation discrimination, dispersion discrimination, area discrimination, shape discrimination, gradation ratio discrimination, length measuring discrimination. A device for inspecting the cleanliness of the top in which a sliver is imaged with an image sensor while being expanded uniformly and fed and the obtained image signals are processed with an image processing device.

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FIG. 1



BACKGROUND OF THE INVENTION

Field of the Invention

5 The present invention relates to a method of and an apparatus for inspecting cleanness of natural fiber tops including wool and cotton for example.

Background Art

10 Normally, inspection of the cleanness of wool tops is executed in accordance with the inspection method prescribed in paragraph 5.6 (ruling the number of nep) and paragraph 5.7 (ruling the number of vegetal impurities) of JIS L-1083. The cleanness is designated by means of the number of detected defects per inspection item present in a predetermined amount of inspected top slivers.

Conventionally, cleanness of wool top sliver is visually inspected. Authorized inspectors visually detect
 15 and classify defects from wool top slivers aligned on an inspection table. Conventionally, a specific continuous inspection method has been executed in accordance with paragraphs 5.6 and 5.7 of JIS L-1083 by thinly and uniformly spreading top slivers in a gill box by about 0.5mm of thickness, and then, sampled top slivers are placed on an inspection table. Inspectors then visually count the number of impurities and defects by irradiating sampled top slivers with light beam permeated through an inspection window. Since
 20 inspectors are obliged to visually count the number of impurities and defects based on the conventional practice, results of the visual inspection are noticeably variable between inspectors.

In order to minimize difference of the inspected results, only those strictly selected inspectors execute inspection of top slivers. Nevertheless, the resultant values are not yet reliable to full extent. Actually, the visually inspected results are merely considered to be relative values of comparison.

25 In order to fully solve this problem, the Applicant of the invention had previously filed an application for a patent as per the Japanese Patent Application No. 62-303563 of 1987, on a method of classifying pillwise defect and vegetal defect from image signal generated from image of top slivers picked up by an image sensor. The proposed inspection method introduces a system for discerning the elongation rate in the field of pillwise defect and vegetal defect. This in turn obliges the proposed method to follow up complex
 30 computation to extract features of defects. As a result of slow speed available for the feature extracting process, the proposed method could not properly be applied to the execution of continuous inspection processes.

Object of the Invention

35 Therefore, the object of the invention is to provide an improved method of and an improved apparatus for precisely and very quickly inspecting cleanness of top slivers.

Disclosure of the Invention

40 The method of inspecting cleanness of sampled top slivers embodied by the invention characteristically comprises a step for generating image signal by picking up image of top slivers with an image sensor and a step for classifying pillwise defect and vegetal defect mixed in the inspected top slivers from the generated image signal based on the discernment of roundness and/or discernment of slenderness.

45 When executing a step for classifying defects, it is desired that discernments of roundness, slenderness, gradation, and dispersion, be introduced. It is also possible for the method embodied by the invention to further classify the pillwise defect into slab, nep, and pinpoint defects based on the area discernment. In addition, the method embodied by the invention can further classify the vegetal defect into trefoil bar and those impurities and bars other than trefoil bar based on the shape discernment. In addition, the method
 50 embodied by the invention can further classify these impurities and bars based on the length discernment.

In the course of classifying those defects cited above, it is particularly effective for the method and apparatus embodied by the invention to execute those sequential processes described below. Initially, the inspecting apparatus determines the range of magnitude of the amount of plural features of respective defects. Next, the inspecting apparatus establishes "fuzzy rules" for classifying all the defects by
 55 integrating two of membership functions to determine the magnitude of probability of being defect against all the defects and also determine adaptability of respective defects to the range of the magnitude of the amount of those features of respective defects. Finally, the inspecting apparatus classifies all the defects based on "fuzzy" reasoning derived from the membership functions and the "fuzzy rules".

The apparatus for inspecting cleanness of top slibers embodied by the invention characteristically comprises the following; a gill box which spreads top slibers into substantially uniform thickness and then externally delivers the spread top slibers, a plurality of rollers which respectively guide the spread top slibers, an illuminating unit which irradiates the spread top slibers with light beam, an image sensor which picks up image of spread top slibers, a contact-free electrostatic removing unit which is installed to a position close to those rollers mentioned above and eliminates static charge from the spread top slibers, and an image processing unit which processes image signal from the image sensor.

BRIEF DESCRIPTION OF THE DRAWINGS

Fig. 1 presents an overall perspective view of the top inspection apparatus embodied by the invention; Fig. 2 is explanatory of the operation of the image processing unit of the top inspection apparatus embodied by the invention; Fig. 3 through Fig. 7 respectively designate the method of processing image of inspected top slibers; Fig. 8 presents an operating flowchart designating steps for processing image of inspected top slibers; Fig. 9 presents an operating flowchart designating the flow of operations for classifying cleanness of the inspected top slibers shown in Fig. 8; Fig. 10 through Fig. 13 are respectively explanatory of a variety of defects and the method of measuring the amount of features of respective defects; Fig. 14 through Fig. 20 respectively and graphically designate membership functions to determine the amount of the features of respective defects; Fig. 21 through Fig. 28 respective and graphically designate membership functions to deal with respective defects; Fig. 29 graphically explains the method of determining adaptability to the amount of features of respective defects; Fig. 30a through 30c respectively explain the method of implementing superimposition; and Fig. 31 presents an operating flowchart of sequential processes for classifying defects of cotton slibers. Fig. 32 shows an image chart designating distribution of depthwise levels of picture elements on the line indicated by character A; Fig. 33 designates the distribution of depthwise levels of pillwise defect; Fig. 34 designates the distribution of depthwise levels of vegital defect;

DESCRIPTION OF THE PREFERRED EMBODIMENTS

Optimal form for embodying the Invention

Referring now more particularly to the accompanying drawings, detail of the method of and apparatus for inspecting cleanness of top sliber according to an embodiment of the invention is described below.

Fig. 1 presents an overall perspective view of the top cleanness inspection apparatus according to an embodiment of the invention. An image sensor (which is substantially a CCD camera) 8 held by a supporting rod 9 is secured to a position above an inspection window 4 of an inspection table 3. The image sensor 8 picks up image of sampled top sliber 1 which is continuously delivered from a gill box 2. An illuminating unit 6 installed below the inspection window 4 emits light beam upward. A roller 10 is installed in front of the image pickup position. The roller 10 prevents the top sliber 1 from superficially incurring fluffing symptom.

An electrostatic preveing unit 30 is provided in the vicinity of a pair of guide rollers 13 and 13a which are respectively installed to the outlet of the gill box 2. The electrostatic preventing unit 30 consists of a rod 31a which is opposite from the upper guide roller 13 and the other rod 31b being opposite from the lower guide roller 13b. A high voltage is delivered to these rods 31a and 31b to permit the needle tips of these rods to respectively execute corona discharge in order to ionize ambient air before eliminating static charge from the charged object in contact-free condition.

Next, image of the top sliber 1 picked up by the image sensor 8 is then converted into an image signal before being delivered to an image processing unit 11. Simultaneously, synchronizing signals needed for continuously picking up image are transmitted to the image processing unit 11 from a synchronizing sensor 12a secured to a measuring roller 12 which is set to a position close to the rear end of the inspection table 3.

The image processing unit 11 extracts a predetermined amount of features, and then, based on the extracted feature data, the image processing unit 11 classifies the inspected top sliber 1 into respective

degrees of cleanness. Finally, the classified result is printed out by a printer 14 so that the result can be made available for the table designating concrete results of the inspection of cleanness.

Fig. 2 presents further detail of the structure of the image processing unit 11. Structurally, the image processing unit 11 is composed of an image input terminal 15, a frame memory 16, a data processor 17, an internal memory 18, an external memory 19, and a system 20 available for printing the table of test results, respectively.

Fig. 3 illustrates a variety of images picked up by the image sensor 8. A variety of impurities and defects present in the sampled top slier 1 are exposed to light beam permeated from the inspection window 4, and then, according to the kind of these impurities and defects, these are converted into black shadows containing depthwise gradation before eventually being caught by the image sensor 8. Next, the image containing depthwise gradation is converted into an image signal "S" (shown in Fig. 4) before eventually being delivered to the image processing unit 11. The image signal shown in Fig. 4 corresponds itself to the image containing depthwise gradation taken on line A. After being received by the image processing unit 11, the image signal "S" is converted by an A/D converter into a multivalent image signal "Sa" (shown in Fig. 5) by providing each picture element with a certain depth. The frame memory 16 stores this multivalent image signal "SA". Next, the data processor 17 makes a comparison between the preset binary level and the multivalent image signal "Sa", and then generates a binary-coded image shown in Fig. 6. Next, the data processor 17 executes a pretreatment, in other words, a labelling operation, in order to secure features present in individual domains of those impurities and defects of the inspected top slier 1 against the binary-coded image ("1" and "0" are respectively the values owned by figure) shown in Fig. 7A, and then, as shown in Fig. 7B, executes a numbering operation. Next, the data processor 17 computes the amount of features present in the numbered individual domains, and then stores the computed result in the internal memory 18. The apparatus embodied by the invention classifies cleanness of the inspected top slier 1 based on the amount of features of defect.

Fig. 8 presents an operating flowchart of those processes needed for discerning cleanness of the inspected top slier 1. The data processor 17 reads the amount of feature from the internal memory 18, and then classifies cleanness. After fully completing the process for classifying cleanness, the data processor 17 rates the cleanness, and then prints the rated cleanness on a table designating the inspected results. If the process for classifying cleanness of the top slier were not yet completed, then, the data processor 17 again reads the amount of feature before eventually classifying the cleanness of the inspected top slier 1.

As is clear from the operating flowchart shown in Fig. 9, the data processor classifies defects of the inspected top slier 1 according to respective check items based on the amount of feature stored in the internal memory 18.

Fig. 9 presents the operating flowchart of the data processor 17 when classifying defects of the inspected wool top slier. While the initial step of classification is underway, the data processor 17 discerns pillwise defect (including slab, nep, pinpoint, and round trefoil bar symptoms) from vegital defect (including impurities and bar symptoms). Roundness and slenderness are effectively made available for composing the amount of feature.

As shown in Figures 10 and 11, concretely, the data processor 17 discerns pillwise defect from vegital defect based on the rating of roundness and slenderness by effectively availing of those features including substantially round shape of the pillwise defect 26 and slender shape of the vegital defect 27. Roundness and slenderness are respectively rated by applying those expressions shown below. The data processor 17 executes classification of defects by comparing the computed values of roundness and slenderness to the predetermined reference value.

$$\text{Roundness} = 4\pi \times (\text{area per picture element}) \times (\text{total number of picture element}) / (\text{peripheral length})^2$$

$$\text{Slenderness} = (\text{area per picture element}) \times (\text{total number of picture element}) / (\text{length of shadow projected against axis X})^2 + (\text{length of shadow projected against axis Y})^2$$

Availing of difference in the light-permeable amount between pills (including slab, nep, and pinpoint) and round trefoil bar, the data processor 17 discerns the pillwise defect from the round trefoil bar based on gradation discernment and dispersion discernment. Concretely, values of gradation and dispersion are computed by applying those expressions shown below. The data processor 17 classifies gradation and dispersion by comparing values of gradation and dispersion to the predetermined reference value.

$$\text{Gradation} = (\text{sum of depth levels}) / (\text{total number of picture element})$$

$$\text{Dispersion} = (1/n) \sum X_i^2 - \bar{x}^2$$

where

- "n" designates the total number of picture elements;
- X_i designates depth levels of picture elements " i "; and
- x designates gradation.

Pillwise defect is classified into slab, nep, and pinpoint according to the size. Therefore, pillwise defect is classified into three categories according to area, which is computed by applying the expression shown below.

$$10 \quad \text{Area} = (\text{area per picture element}) \times (\text{total number of picture element})$$

The data processor 17 classifies pillwise defect into slab, nep, and pinpoint in order of the size by making a comparison between the computed area value and the predetermined reference value. The data processor 17 further classifies the vegital defect into impurities, bar, and trefoil bar, by discerning these shapes. Concretely, shape is computed by applying the expression shown below.

$$\text{Shape} = (\text{peripheral length}) / \sqrt{(\text{length of shadow projected against axis X})^2 + (\text{length of shadow projected against axis Y})^2}$$

Fig. 12 graphically designates defect 28 caused by impurities. Assume that the defect 28 contains about 57 of peripheral length, 20 of the length "e" of shadow projected against axis X, and 20 of the length "f" of shadow projected against axis Y, then the shape of this defect 28 is substantially doubled as per the solution $57 / \sqrt{20^2 + 20^2}$ of the above expression.

Fig. 13 graphically designates defect 29 caused by trefoil bar. Assume that this defect 29 contains about 97 of peripheral length, 20 of the length "e" of shadow projected against axis X, and 15 of the length "f" of shadow projected against axis Y, then, the shape of this defect 29 is quadrupled as per the solution $97 / \sqrt{20^2 + 15^2}$ of the above expression. Concretely, the data processor 17 discernibly classifies defects into the one caused by the impurities and the other one caused by trefoil bar by making comparison between the above numerical values and the predetermined reference value. The data processor 17 discerns impurities from bar based on the measurement of length. Length is computed by applying the expression shown below.

$$\text{Length} = \sqrt{(\text{length of shadow projected against axis X})^2 + (\text{length of shadow projected against axis Y})^2}$$

Incidentally, the defect 20 caused by the impurities shown in Fig. 12 has about 28 of the length value as per the solution $\sqrt{20^2 + 20^2}$ of the above expression. The data processor 17 discerns this defect for classification by comparing it to the predetermined reference value. The trefoil bar cited above is eventually added to the category of bar.

In accordance with the operating flowchart shown in Fig. 8 and based on the result of classifying all defects, the data processor 17 executes the rating of cleanness of the inspected wool top sliber. The rated cleanness is effectively made available for composing quality standard of the top. The established quality standard is also effectively made available for composing basic data when spinning wool top sliber into yarns in the following processes.

Next, the method of classifying those defects by applying "fuzzy" reasoning is described below.

Initially, based on the definition of membership function, the data processor 17 establishes a specific range of those adjective expressions including "big", "intermediate", and "small" which are respectively applicable to the amount of features of defect including roundness, slenderness, area value, shape value, gradation value, length value, and dispersion value, and the other specific range of those adjective expressions including "high", "normal", and "low" for designating probability of being defect. The definition of membership function may be established in accordance with the judgement of those who are extremely skilled in the related art.

Next, using the range which is applicable to those adjective expressions like "big", "intermediate", and "small" cited above, the data processor 17 defines "fuzzy rule" for designating the range of the amount of features of defect in terms of the relationship between the combination and the probability of being specific defect, in other words, the data processor 17 defines the "fuzzy rule" which designates the relationship between the conditional terms like "if it were ---" and the conclusive terms like "probability of ---" for example.

Table 1 concretely designates the membership function and the "fuzzy rule" cited above. For example,

if there were substantial roundness and slenderness, substantial area value, and in addition, more than intermediate rating of gradation, then, it is highly probable that the defect is caused by presence of slab, and therefore, there is less probability that the defect is caused by presence of pinpoint (p.p.) and noise. Neither high probability nor low probability of being nep, or trefoil bar, or bar, is present in the above defect, but the probability is rated to be normal. In this way, relative to the conditional section shown to the left of Table 1, the conclusion section is shown to the right.

Next, procedure for executing the "fuzzy reasoning" based on the above membership function and the "fuzzy rule" is described below.

Initially, input data is converted in correspondence with the amounts of respective features of a single defect. Next, adaptability to those adjective terms like "big", "intermediate", or "small", is computed from the membership function. Of those amounts of features of defect, if the objective shape were substantially circular, then the roundness is rated to be very close to 1. As mentioned earlier, the shape of pillwise defect is closer to circle than that of vegital defect, and thus, the difference between the pillwise defect and the vegital defect can be expressed by means of the rating of roundness. The rating of roundness is in a range from 0 to 1. This range is converted into 0 through 100(%). If the objective shaper were thinner, then the rating of slenderness approximates 0. Since the vegital defect has shape thinner than the pillwise defect, the difference between both defects can be designated. The rating of slenderness is in a range from 0 to 0.5. This range is converted into 0 through 100(%). Substantially, the gradation value represents the mean value of the depth of objects. Since more amount of light permeates the pillwise defect than the case of permeation through the vegital defect, the pillwise defect generate an image containing high-level density. The gradation value is in a range from 145 to 170. This range is also converted into 0 through 100(%). The dispersion value represents the unevenness of the depth of objects. Since the amount of light permeating plant and sliber differs, availing of substantial dispersion of depth level, precision in the classification of trefoil bar can be promoted even when sliber tangles trefoil bar. The dispersion value is in a range from 0 to 25(%). Like the above cases, this range is converted into 0 through 100(%). The shape value is incremental relative to the increase of shapes of objects by plural number. Since trefoil bar has shape being more complex than that of impurities and bars, availing of this complex shape, difference between trefoil bar and impurities/bar can effectively be designated. The shape value is in a range from 0 to 5. This range is also converted into 0 through 100(%). The area value designates areas of objects, in which pinpoint (p.p.), nep, and slab, are respectively classified according to own size, and thus, difference between these can be designated by means of area value. The area value is in a range from 0 to 100(%). This range is also converted into 0 through 100(%). The length value designates the length of objects. Availing of the length value, the data processor 17 discerns bar (having a minimum of 10mm of length) from impurities (having a minimum of 3mm and a maximum of 10mm of length). The length value is in a range from 0 to 10. This range is also converted into 0 through 100(%).

Fig. 14 through Fig. 20 graphically illustrate the membership function of roundness, slenderness, area value, shape value, gradation value, length value, and dispersion value. The label PS shown in these graphic charts designates "small". The label PM designates "intermediate". The label PB designates "big". The label VS designates "very small". The label VB designates "very big". The label PM< shown in the gradation value designates "being greater than intermediate"

TABLE 1 Contents of Fuzzy Rule

No.	Category of defect		Conditional section		Roundness	
5	<u>Slenderness</u>	<u>Area value</u>	<u>Shape value</u>	<u>Gradation value</u>		
	<u>Length value</u>	<u>Dispersion value</u>				
10	<u>Pillwise defect</u>	Big	Big	Big	Greater than intermediate	
	Big	Big	Intermediate	Big	Big	Small Big
	Big	Big	Small	Big	Big	Very small Very big
15	Big	Big	Very small	Very big		
	Big	Big	Very small	Big	Big	Big Very small Big
20	Big	Big	Small	Intermediate	Big	Big Very small
	Intermediate	Very big	Small			
	<u>Vegital defect</u>	Big	Big	Very small	Small	Small Small
25	Big	Small	Small	Small	Small	Big Small Small
	Small	Small	Small	Intermediate	Small	Big Very big
30	Intermediate	Big				

<u>Conclusion section</u> (probability of being defect of								
<u>Slab</u>	<u>Nep</u>	<u>Pinpoint</u>	<u>Trefoil bar</u>	<u>Bar</u>	<u>Impurities</u>	<u>Noise</u>		
High	Normal	Low	Normal	Normal	Low	Normal	High	Normal
Low	Low	Low	Normal	High	Low	Normal	Low	Low High
Low	Normal	Normal	Low	Normal	Normal	Normal	Normal	High

TABLE 2

<u>Amount of feature</u>	<u>Example of input data</u>	<u>Input value</u>
<u>Converted value (%)</u>	<u>Adaptability to words</u>	$(0 \leq \text{Adaptability} \leq 10)$
<u>1. Roundness</u>	<u>2. Slenderness</u>	<u>3. Area value</u>
<u>4. Shape value</u>	<u>5. Gradation value</u>	<u>6. Length value</u>
Very small	Small	
Intermediate	Big	Very big
		Very big

TABLE 3 Adaptability of Rule (Rule No. 11 for example)

<u>Conditional section</u>	<u>Adaptability</u>	<u>Conclusion section</u>
<u>Adaptability</u>	1. Roundness is small	Slab
	2. Slenderness is small	Nep
	3. Area value	
	4. Shape value is small	Pillwise noise
	5. Gradation value is small	Probability of being trefoil bar is low.
	6. Length value is big	Probability of being bar is high.
		Probability of being impurities is normal.
		Probability of being vegital noise is low.

Adaptability of Rule No. 11 (Introduces minimum value of
adaptability in the Conditional
section)

TABLE 4 Results of defect classification

Defect detection rate	Defect discerning rate
(A/C X 100%)	(B/D X 100%)

Total number of data Rate of achieving correct answer
Via fuzzy reasoning Via conventional practice

- A. Number of data which correctly discerned the defects from defect-free objects
- B. Number of data which correctly discerned all defects from each other
- C. Total number of data dealt by the data processor
- D. Number of data which contained all the designated defects among those which are dealt by the data processor

Fig. 21 through Fig. 28 graphically illustrate the membership function of slab, nep, pinpoint, pillwise noise, trefoil bar, bar, impurities, and vegital noise, which are respectively determined to be defect. The label PS designating probability indicates "low". The label PM indicates "normal". The other label PB indicates "high".

On receipt of those input data designating the roundness and other check items from the image processing signal, the data processor 17 computes adaptability of those adjective words "big", "intermediate", and "small" contained in the input values by applying those values converted from the input data. For example, if the input value of length value contained in an input data designating certain defect like bar were 9.54 and the converted value 95.4, then, as shown in Fig. 29 designating the membership function of the length value, the data processor 17 figures out the adaptability to the term "big" to be 8.11, adaptability to the term "intermediate" to be 0.73, and the adaptability to the term "small" to be zero (the PS is free of intersecting point) from the vertical coordinate at the intersecting point of the vertical line of the input value 9.54 against inclined lines PB, PM, and PS. In the same way, the data processor 17 computes the adaptability to the amount of other features. Table 2 presents an example in which the data processor 17 computes adaptability to those adjective words in the course of dealing with a variety of defects containing 9.54 of the input length value, 0.24 of the input roundness value, 0.09 of the input slenderness value, 326.00 of the area value, 2.19 of the shape value, and 135.10 of the gradation value.

Next, the data processor 17 computes adaptability to respective rules by applying the adaptability to those adjective words designating the amount of features generated by applying those processes described above. For example, Table 3 designates the adaptability to the Rule No. 11 shown in Table 1. Concretely, as shown in Table 3, adaptability of the conclusion section (including the first through seventh conclusions) of the conclusion section of the Rule 11 is arranged to be the minimum value of adaptability in the conditional section shown to the left of Table 3. In this instance, the minimum value is 6.24.

After determining the adaptability of the whole rules as described above, using the determined adaptability, the data processor 17 then executes an operation for computing minimum membership function of the corresponding conclusion section. Next, the data processor 17 executes an operation for computing maximum value to determine whether the probability of falling under respective defects is "high", or "normal", or "low". For example, by referring to the bar-defect determined by the fifth conclusion and based on the adaptability and the computed result, the data processor 17 computes minimum values of PB according to the adaptability value 6.24 of the Rule No. 11, PM according to the adaptability value 0.73 of the Rule No. 12, and PB according to the adaptability value 4.76 of the Rule No. 10. After synthesizing the computed values the data processor 17 establishes the relationship between the rule adaptability and the probability output value as shown in Fig. 30a. Likewise, Fig. 30a designates the synthesized result of the rule adaptability to defective impurities and Fig. 30b the synthesized result of the rule adaptability to vegital noise.

After completing those operations for computing maximum values, based on the yielded membership function, the data processor 17 computes confirmed values of probability of being defect. There are a variety of methods available for computing confirmed values of the probability including application of the center of gravity, center-computing method, height-computing method, and area-computing method. The data processor 17 computes confirmed values of the probability of being defect by applying more than one of those methods cited above. When computing the probability output values (confirmed values) of bar, impurities, and vegital noise based on the application of the center of gravity for example, the confirmed

value of the bar defect is determined to be 53.4, the confirmed value of the impurities to be 38.6, and the confirmed value of the vegital noise to be 30.5. Based on these confirmed values, the data processor 17 concludes that the defect is substantially caused by presence of bar. In the event that the computation of confirmed value yields certain values being equal to each other in two kinds of defect then confirmable answer may be yielded by applying those different computing methods before eventually determining the confirmed value by accepting a majority decision.

Table 4 presents the results of the actually executed top inspection by applying the defect-classifying method via "fuzzy" reasoning. Table 4 also presents the result of inspecting defects based on a conventional method. As is clear from Table 4, although there is no substantial difference in the rate of detecting defect between those results yielded from the conventional method and the method embodied by the invention, the comparative result proves that the rate of yielding correct answer from the defect discernment based on the "fuzzy" reasoning introduced to the invented method has resulted in noticeable improvement.

As is clear from the above description, since the method embodied by the invention precisely classifies pillwise defect and vegital defect based on the roundness discernment and/or slenderness discernment, amount of data to be processed in the course of extracting features of defect sharply decreases. This in turn permits the method and apparatus embodied by the invention to significantly accelerate defect inspection with improved data processing efficiency.

Furthermore, as a result of the addition of a gradation discerning process and a dispersion discerning process to the roundness and slender discernments, the method and the apparatus embodied by the invention can precisely discern trefoil bar from slab, thus significantly promoting accuracy of the inspected results.

The method and the apparatus embodied by the invention precisely classify the pillwise defect further into those defects including slab, nep, and pinpoint, by applying area discernment, and yet, further classify the vegital defect into trefoil bar and those defects other than the trefoil bar like impurities and bar by applying shape discernment. The method and the apparatus embodied by the invention further classify impurities from bar by applying length discernment. This in turn permits the method and the apparatus embodied by the invention to finely classify defects per check item, thus offering practical advantage.

Furthermore, when classifying those defects based on the "fuzzy" reasoning, the method and the apparatus embodied by the invention do not respectively identify those defects based on an explicit threshold value, but the invented system solely reasons those defects by applying ambiguously defined domain, and therefore, the method and the apparatus embodied by the invention can achieve an overall result of judgement close to human judgement. This in turn permits the method and the apparatus embodied by the invention to very precisely discern those confusing plural defects like those which are actually present in the top slier.

Fig. 31 presents an operating flowchart available for discerning defects of cotton slier. The data processor 17 initially executes the dispersion discernment to disperse defects and noise, and then, discerns defects from each other based on the comparison of gradation. The data processor 17 discerns defects based on the comparison of gradation by checking the periphery of a specific area detected by means of binary code, and then, the periphery of the detected area and the gradation are compared to each other. The defect containing substantial gradation rate is identified to be noise. The defect containing negligible gradation rate is identified to be vegital mixture, whereas the defect containing modest gradation rate is identified to be pillwise defect. There are six kinds of defect inherent in cotton slier including large, modest, and small vegital refuse, and large, modest, and small pillwise nep. However, there is little shapewise difference between vegital refuse and pillwise nep, and yet, unlike wool, there is no difference of size between these. Because of this, when executing a conventional method of classifying the cleanness of wool top, actually, refuse and nep are often incorrectly discerned from each other.

To solve this problem, it is quite effective for the defect inspecting system to introduce the method of discerning defects based on the comparison of gradation described above. Concretely, since pillwise defect is generated by tangled fibers, variation of depthwise level from the periphery of defect to the defective domain does not sharply occur unlike the case of vegital defect. For example, the image chart shown in Fig. 32 designates distribution of depthwise levels of picture elements on the line indicated by character A. Fig. 33 designates the distribution of depthwise levels of pillwise defect. Fig. 34 designates the distribution of depthwise levels of vegital defect. Based on these charts, the following expressions are established.

Gradation = (sum of depthwise levels shown in shadowed domain)/(the number of picture elements shown in the shadowed domain)

Peripheral gradation ratio = (gradation)/[(sum of depthwise levels in the dotted domains)]/[(the number of

picture elements in the dotted domains)]

where those values in brackets [] respectively designate gradation of domains in the periphery of defect.

As is clear from the above expressions, the less the difference of gradation between the defective domain and the peripheral domain, the closer the peripheral gradation ratio to value 1. Therefore, peripheral gradation ratio of pillwise defect exceeds that of vegital defect. The method and the apparatus embodied by the invention can precisely discern the pillwise defect from the vegital drect based on those processes described above. When computing the gradation ratio, the apparatus embodied by the invention initially establishes proper membership function, and then correctly discerns the pillwise defect from the vegital defect based on the "fuzzy" reasoning.

Industrial Applicability of the Invention

As is clear from the above description, the method and the apparatus for inspecting cleanness of top sliber embodied by the invention precisely and quickly analyzes cleanness of natural fibers like wool and/or cotton with perfect uniformity.

TABLE 1 Contents of Fuzzy Rule

No	Category of defect	Conditional section							Conclusion Section (probability of being defect of ...)						
		Roundness	Slenderness	Area value	Shape value	Gradation value	Length value	Dispersion value	Slab	Nep	Pin point	Trefollicar	Bar	Impurities	Nois
1	Pillwise defect	Big	Big	Big		Greater than intermediate			High	Normal	Low	Normal	Normal		Low
2		Big	Big	Intermediate		Big			Normal	High	Normal	Low			Low
3		Big	Big	Small		Big			Low	Normal	High	Low			Normal
4		Big	Big	Very Small		Very Big			Low	Low	Low	Low			High
5		Big	Big	Very Small		Big			Low	Normal	Normal	Low			Normal
6		Big	Big	Small		Intermediate					Normal				
7		Big	Big	Very Small		Intermediate					Normal				
8				Very Big				Small	High						
9	Vegetal defect	Big	Big	Very small		Small						Low	Low	Normal	High
10		Small	Small		Big	Small						High	Low	Low	Low
11		Small	Small		Small	Small	Big					Low	High	Normal	Low
12		Small	Small		Small	Small	Intermediate					Low	Normal	High	Normal
13		Small	Small	Small	Small	Small	Small					Low	Low	Normal	High
14		Small	Small	Very small	Small	Small	Small					Low	Low	Normal	High
15		Big	Big	Very Big		Intermediate		Big				High			

TABLE 2

Amount of feature	Example of input data		Adaptability to words ($0 \leq \text{Adaptability} \leq 10$)
	Input value	Converted value(%)	
Roundness	0. 2 4	8. 0	small:10.00 Intermediate:0.00 Big:0.00
Slenderness	0. 0 9	1 8. 0	small:10.00 Big:0.00
Area value	3 2 6. 0 0	6 5. 2	very small:0.00 small: 0.00 Intermediate:5.80 Big:0.00 very big:0.00
Shape value	2. 1 9	4 3. 8	small: 6.24 Big:4.76
Gradation value	1 3 5. 1 0	0. 0	small:10.00 Intermediate:0.00 Big:0.00 very big:0.00
Length value	9. 5 4	9 5. 4	small: 0.00 Intermediate:0.73 Big:8.11

TABLE 3 Adaptability of Rule (Rule No.11 for example)

No.11	Conditional section	Adaptability	Conclusion section	Adaptability
1	Roundness is small	: 10.00	Slab	
2	Slenderness is small	: 10.00	Nep	
3	Area value		Pin point	
4	Shape value is small	: 5.24	Pillwise noise	
5	Gradation value is small	: 10.00	Probability of being trefoil bar is low.	: 6.24
6	Length value is big	: 8.11	Probability of being bar is high.	: 6.24
7			Probability of being impurities is normal.	: 6.24
8			Probability of being vegetal noise is low.	: 6.24
Adaptability of Rule No.11 6.24 (Introduces minimum value of adaptability in the Conditional section)				

TABLE 4 Results of defect classification

	Defect detection rate ($A/C \times 100\%$)	Defect discerning rate ($B/D \times 100\%$)
Total number of data	705 (=C)	141 (=D)
Rate of achieving correct answer	Via fuzzy reasoning	81.6%
	Via conventional practice	62.4%

- A. Number of data which correctly discerned the defects from defect-free objects
- B. Number of data which correctly discerned all defects from each other
- C. Total number of data dealt by the data processor
- D. Number of data which contained all the designated defects among those which are dealt by the data processor

Claims

1. A method of inspecting cleanness of top sliber comprising the sequential steps of;
a step for generating image signal by picking up image of top sliber with an image sensor; and
a step for discernibly classifying defect into pillwise defect and vegital defect mixed in said top sliber based on said generated image signal via roundness discernment and/or slenderness discernment.
2. A method of inspecting cleanness of top sliber comprising the sequential steps of;
a step for generating image signal by picking up image of top sliber with an image sensor; and
a step for discernibly classifying defect into pillwise defect and vegital defect mixed in said top sliber based on said generated image signal via roundness discernment, slenderness discernment, gradation discernment, and dispersion discernment.
3. A method of inspecting cleanness of top sliber comprising the sequential steps of;
a step for generating image signal by picking up image of top sliber with an image sensor;

a step for discernibly classifying defect into pillwise defect and vegital defect mixed in said top sliber based on said generated image signal via roundness discernment, slenderness discernment, gradation discernment, and dispersion discernment; and

a step for discernibly classifying said pillwise defect further into those defects including slab, nep, and pinpoint, by applying area discernment and said classified vegital defect further into trefoil bar and those defects other than said trefoil bar like impurities and bar by applying shape discernment, and yet, further classifies said impurities and bar by applying length discernment.

4. A method of inspecting cleanness of top sliber comprising the sequential steps of;

a step for generating image signal by picking up image of top sliber with an image sensor; and

a step for discernibly classifying defect into pillwise defect and vegital defect mixed in said top sliber based on said generated image signal;

wherein in the course of classifying said defects, said method establishes membership function (solely available for conditional section) for dealing with features of classified defects and the other membership function (solely available for conclusion section) for determining probability of being defect against said classified defects, wherein said method also establishes fuzzy rule by applying said conclusion section determining probability of being defect against those classified defects; and wherein said method further classifies defects based on fuzzy reasoning derived from said membership function solely available for said conditional section, the other membership function solely available for conclusion section, and said fuzzy rule.

5. A method of inspecting cleanness of top sliber comprising the sequential steps of;

a step for generating image signal by picking up image of said top sliber with an image sensor;

a step for classifying defect into pillwise defect and vegital defect mixed in said top sliber based on said generated image signal by applying gradation discernment and/or gradation ratio discernment; and

a step for further classifying said pillwise defect and vegital defect into respective defects by applying length discernment.

6. An apparatus for inspecting cleanness of top sliber comprising;

a gill box which uniformly spreads top sliber and then externally delivers it;

a plurality of guide rolls which respectively guide said spread top sliber;

an illuminating unit which irradiates said spread top sliber with light beam;

an image sensor which picks up image of said spread top sliber;

a contact-free electrostatic preventing unit which is installed in the vicinity of said rollers and eliminates static charge from said spread top sliber; and

an image processing unit which processes image signal from said image sensor.

FIG. 1

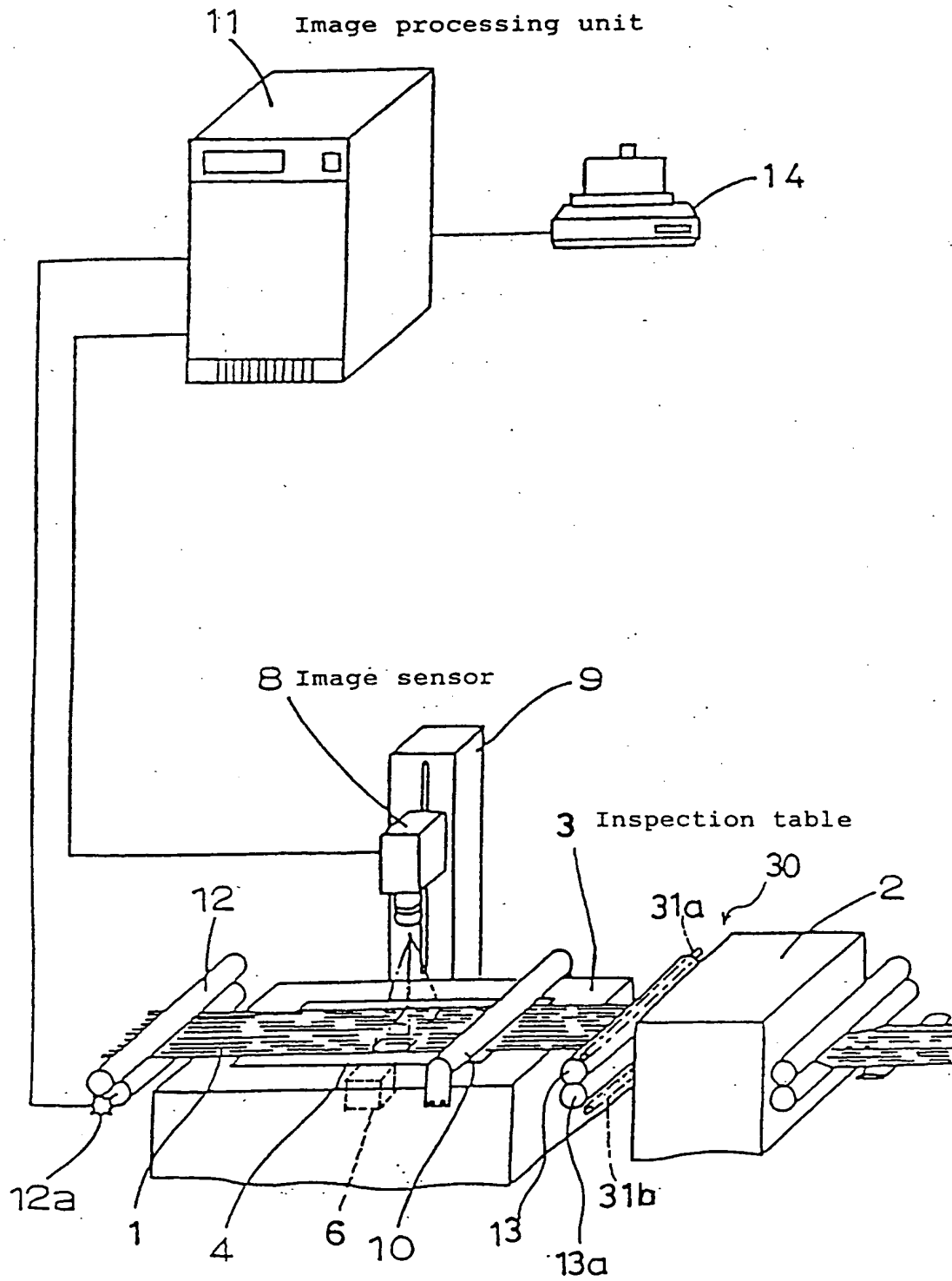


FIG. 2

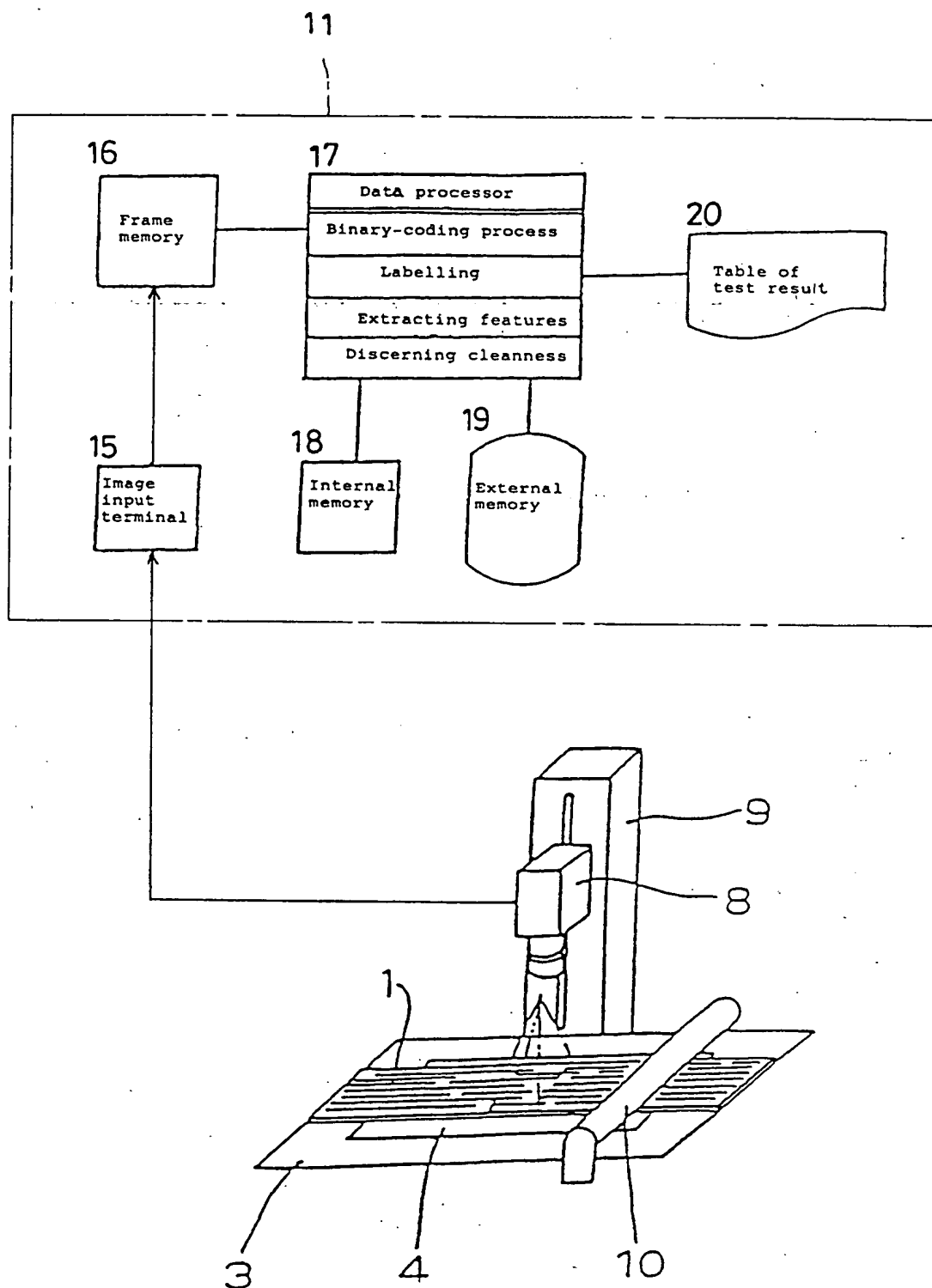


FIG.3

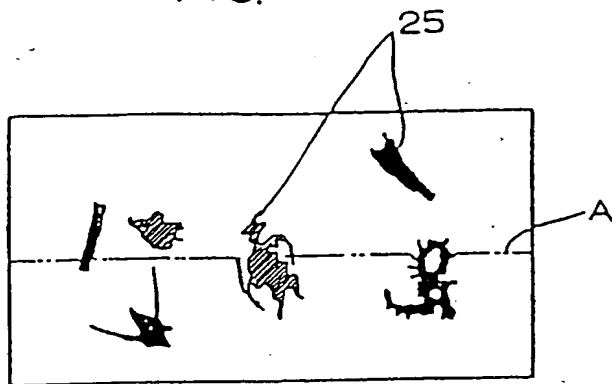


FIG. 4

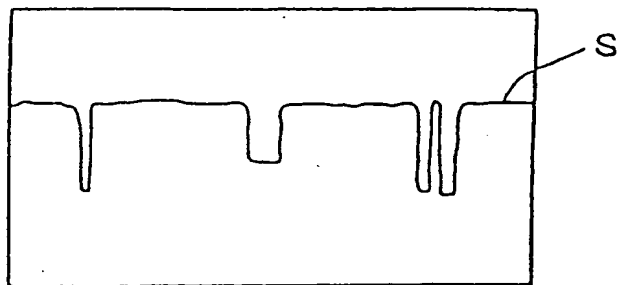


FIG. 5

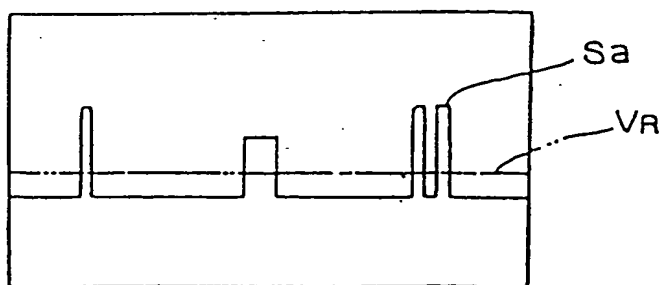


FIG. 6

00001000000000111000000010100000

FIG. 7

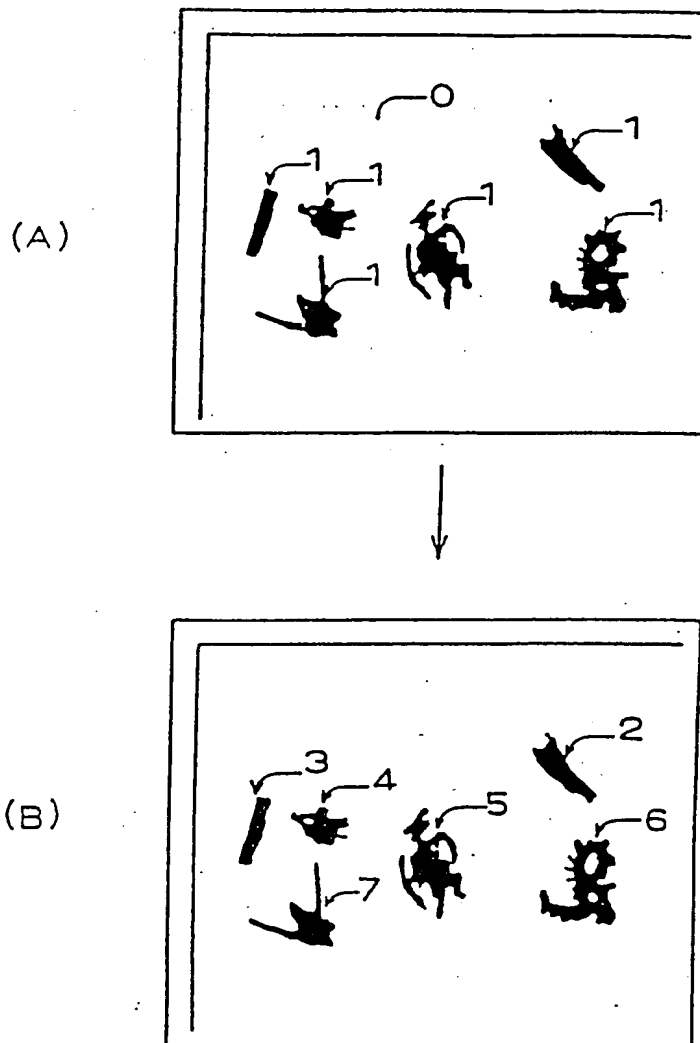


FIG. 8

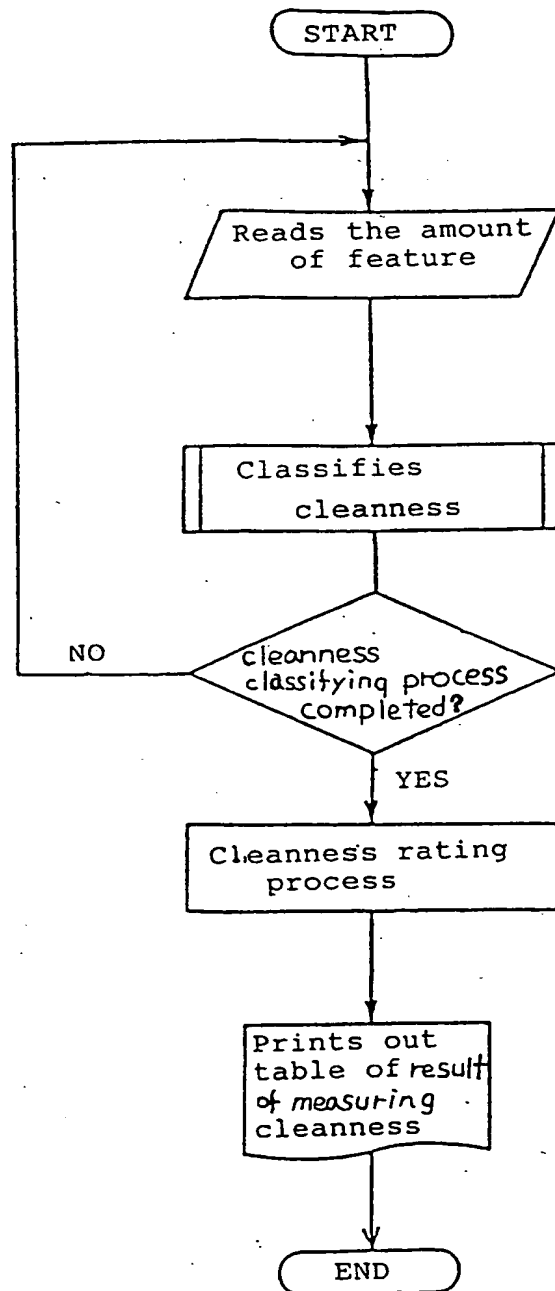


FIG.9

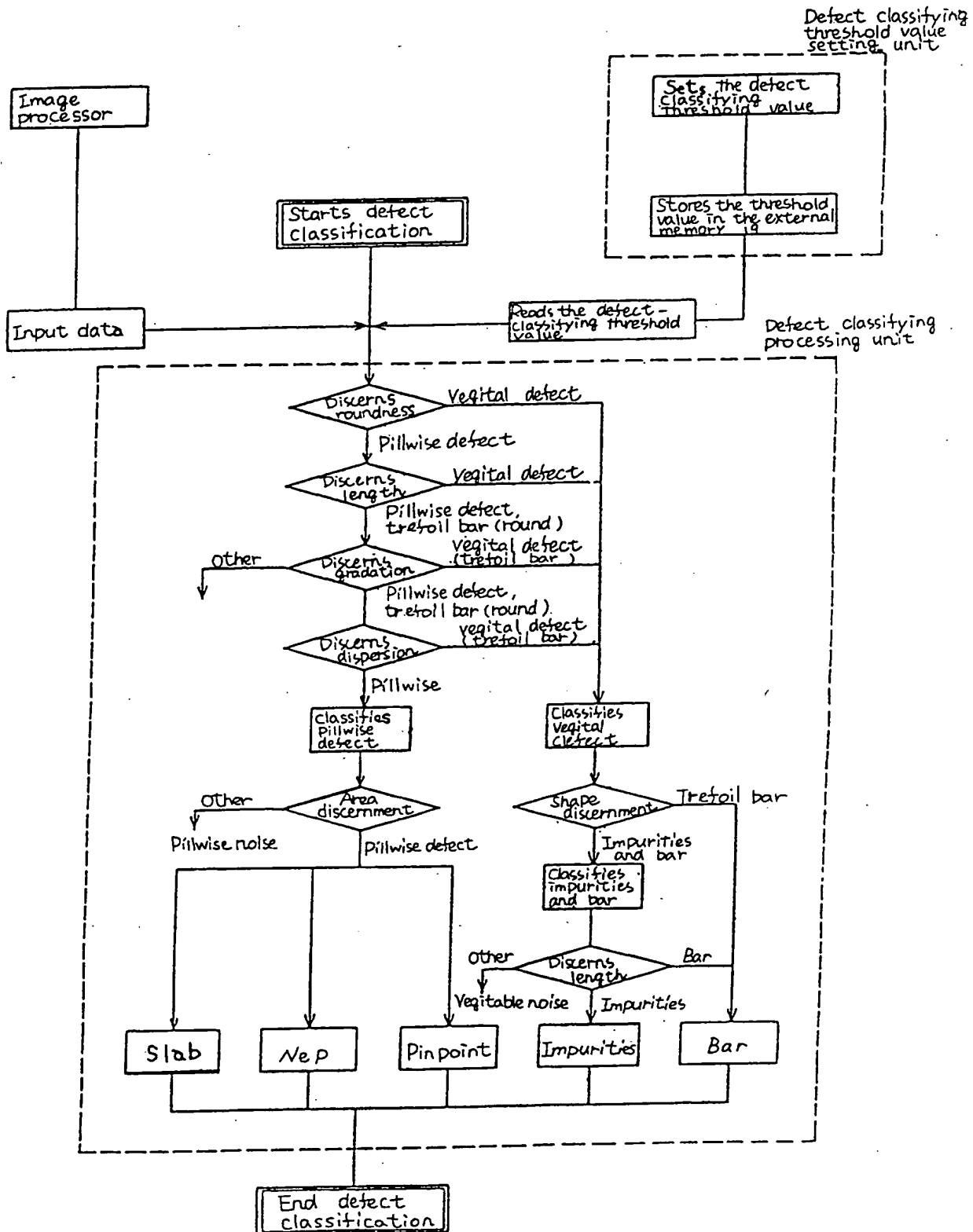


FIG.10

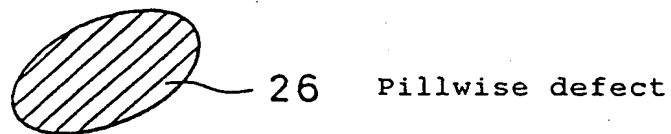


FIG.11

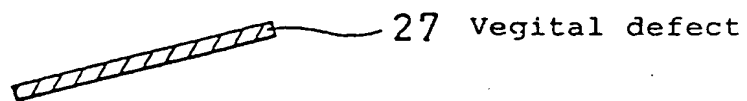


FIG.12

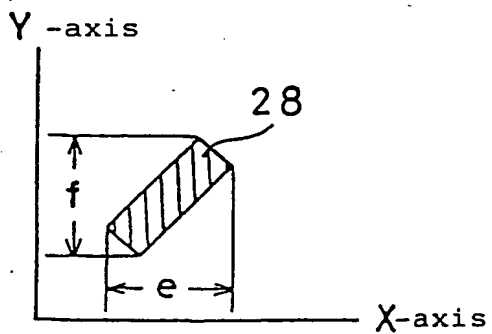


FIG.13

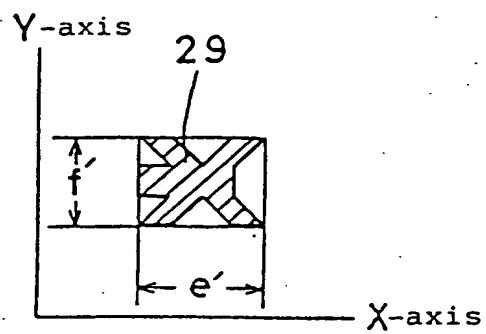


FIG.14

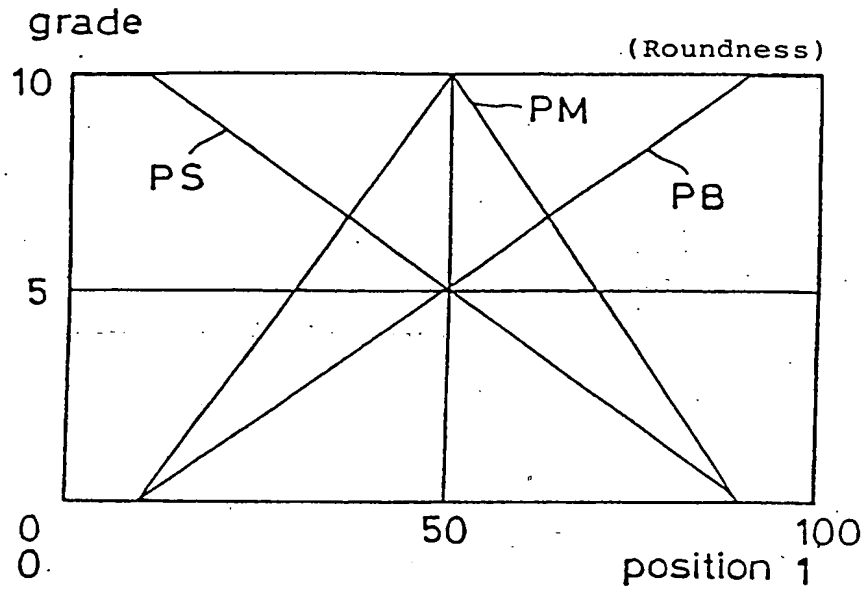


FIG.15

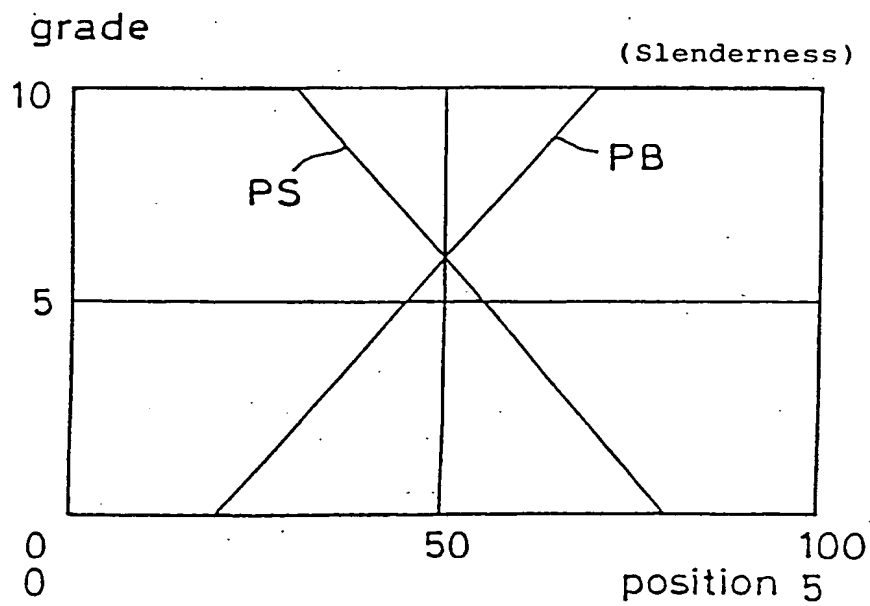


FIG.16

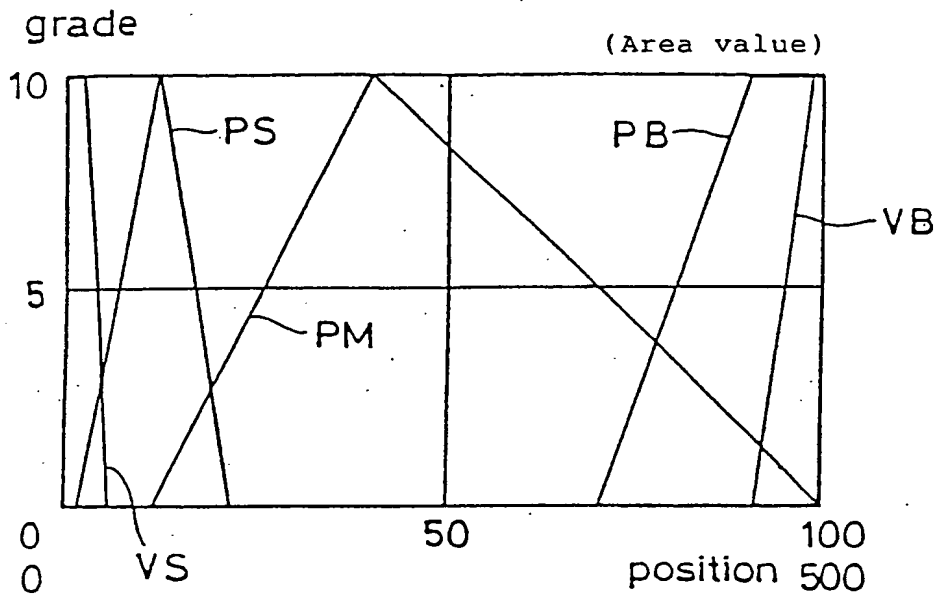


FIG.17

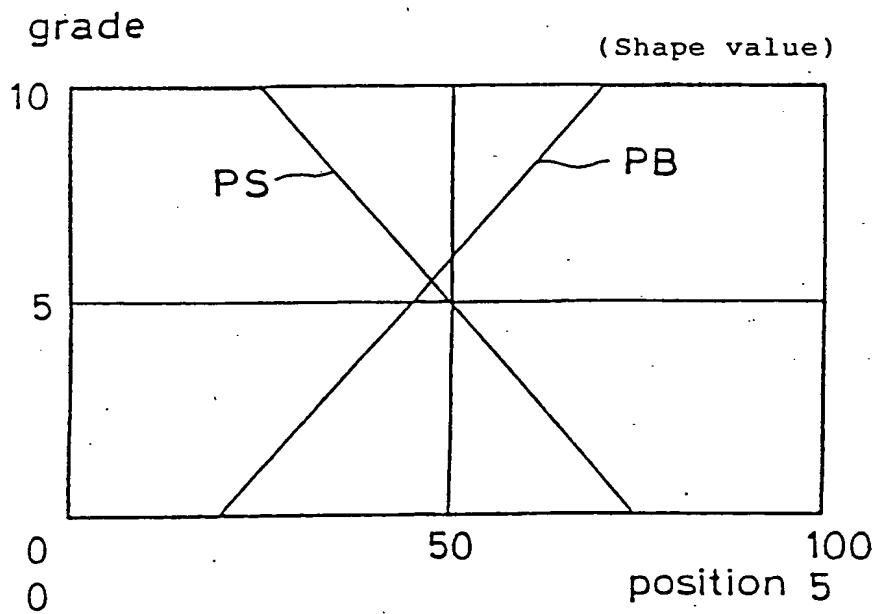


FIG.18

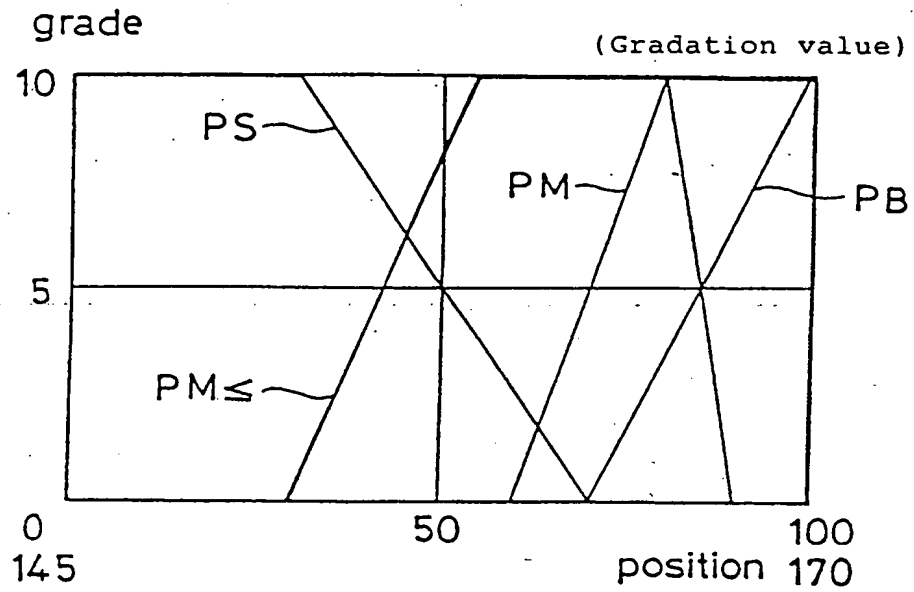


FIG.19

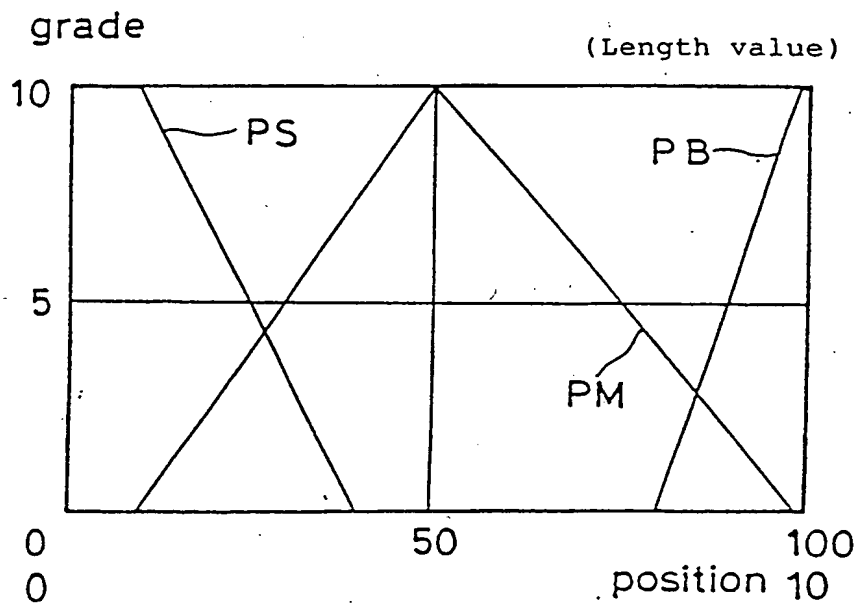


FIG.20

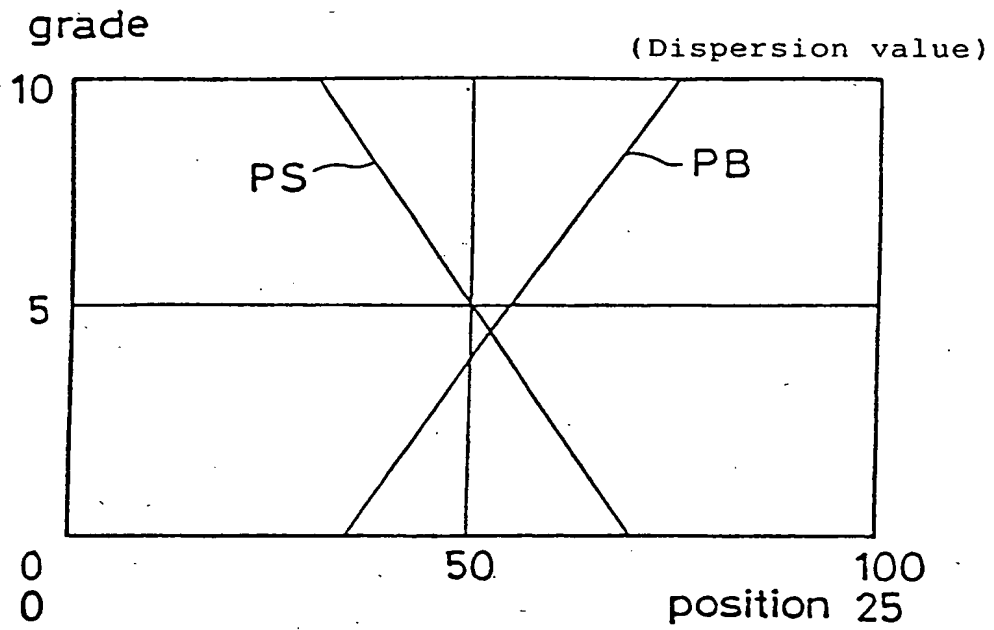


FIG.21

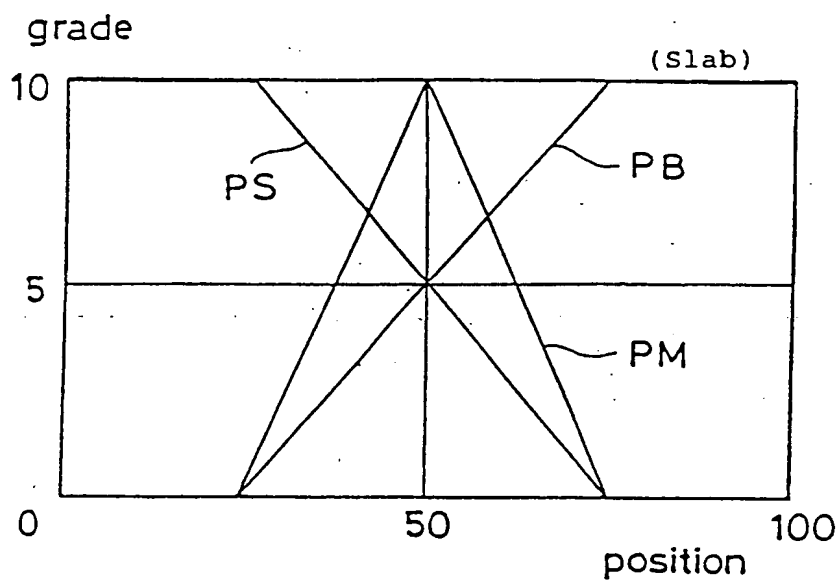


FIG.22

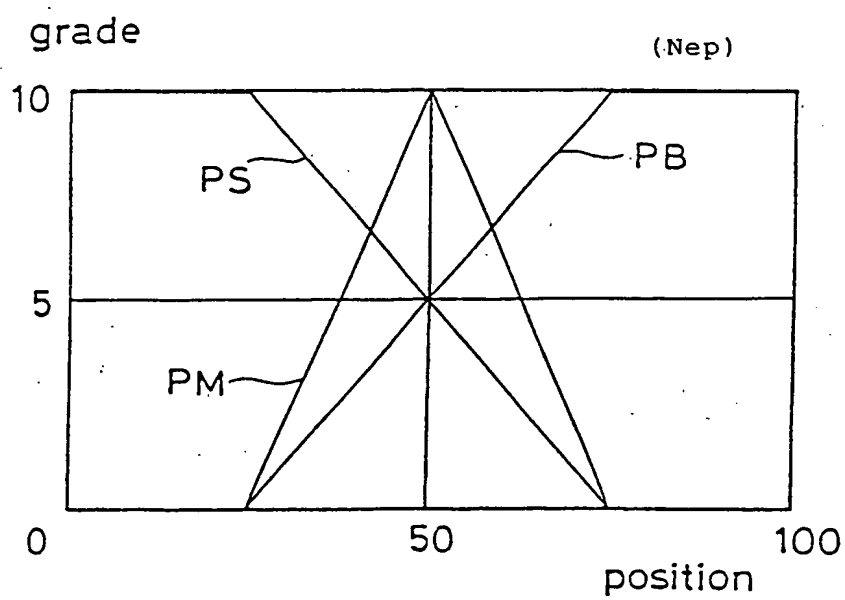


FIG. 23

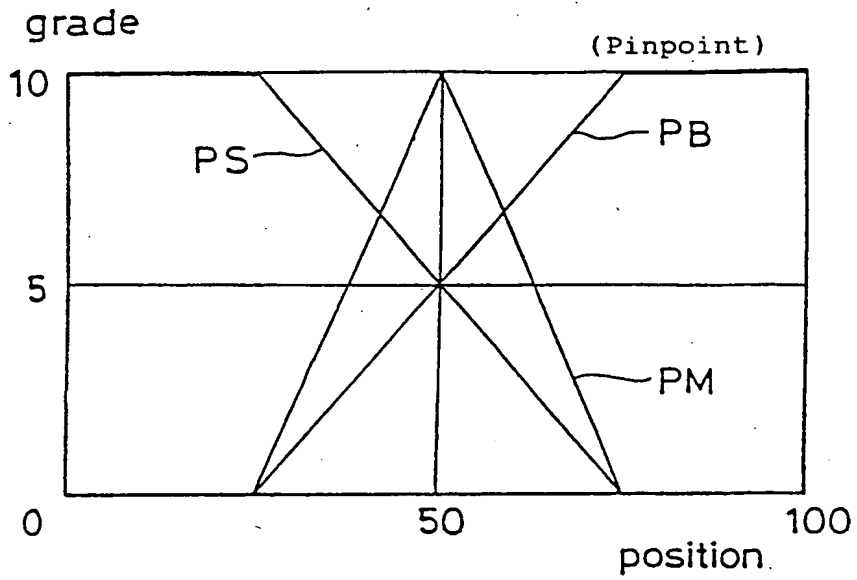


FIG. 24

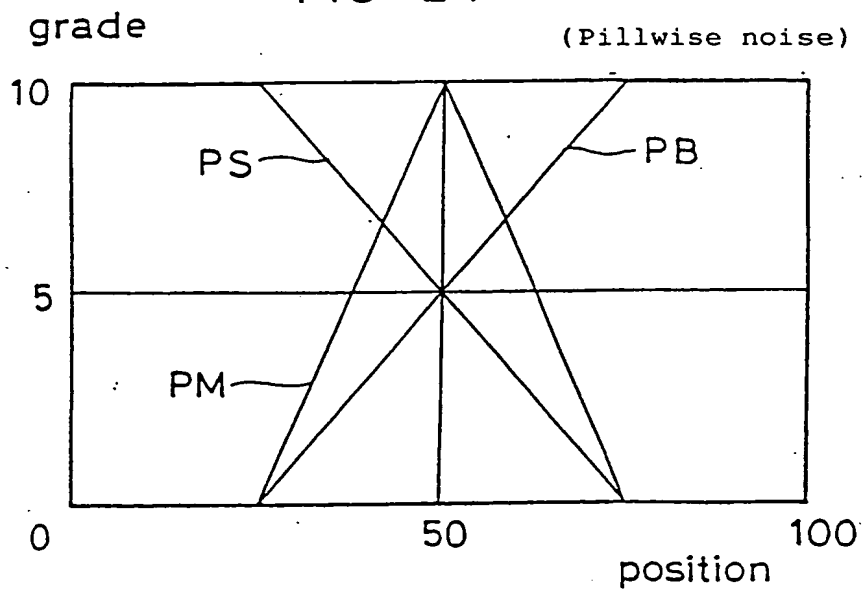


FIG.25

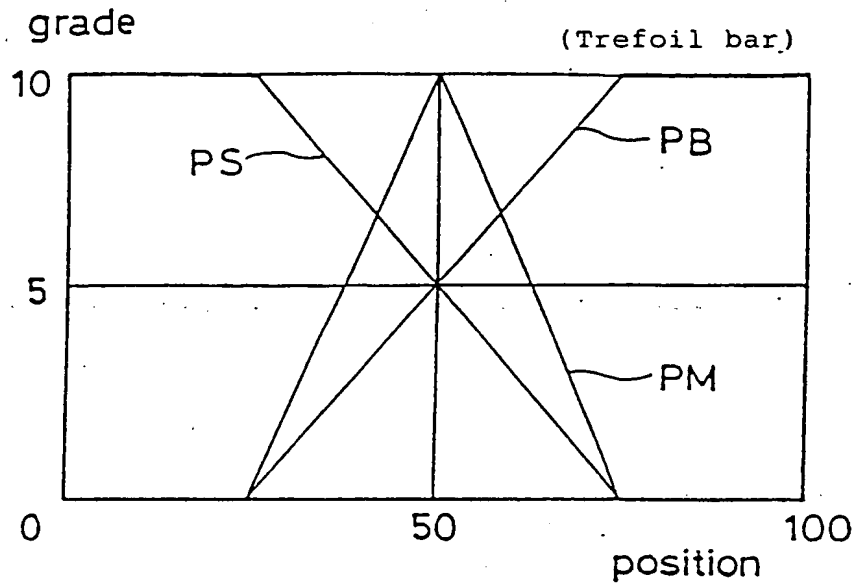


FIG.26

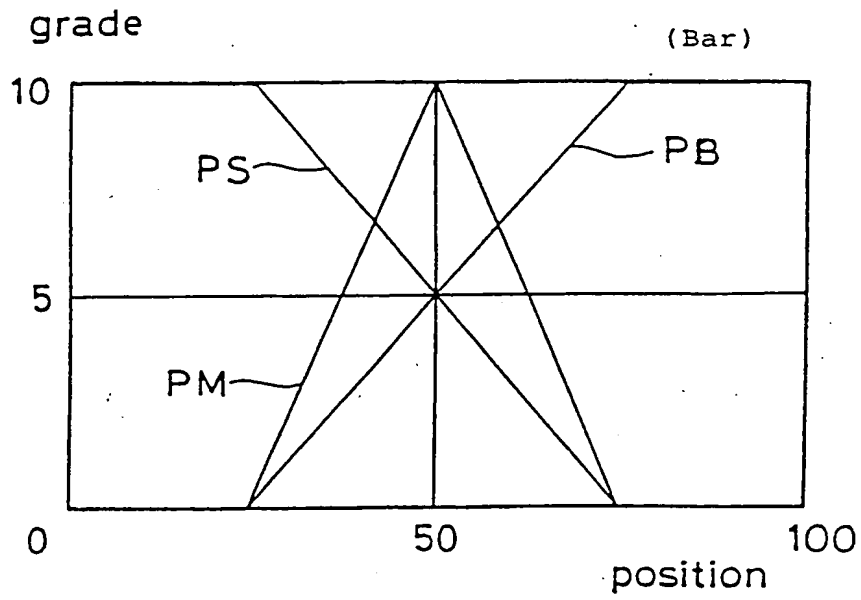


FIG.27

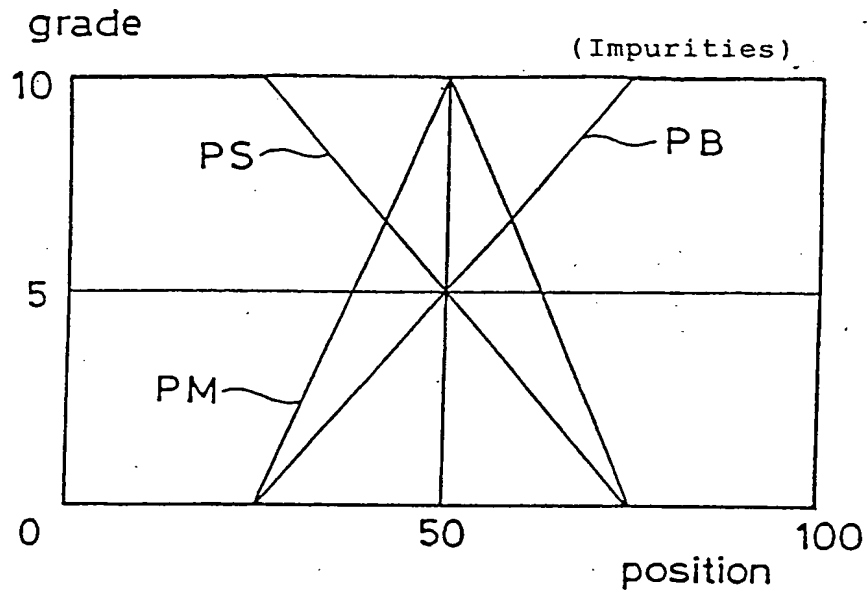


FIG.28

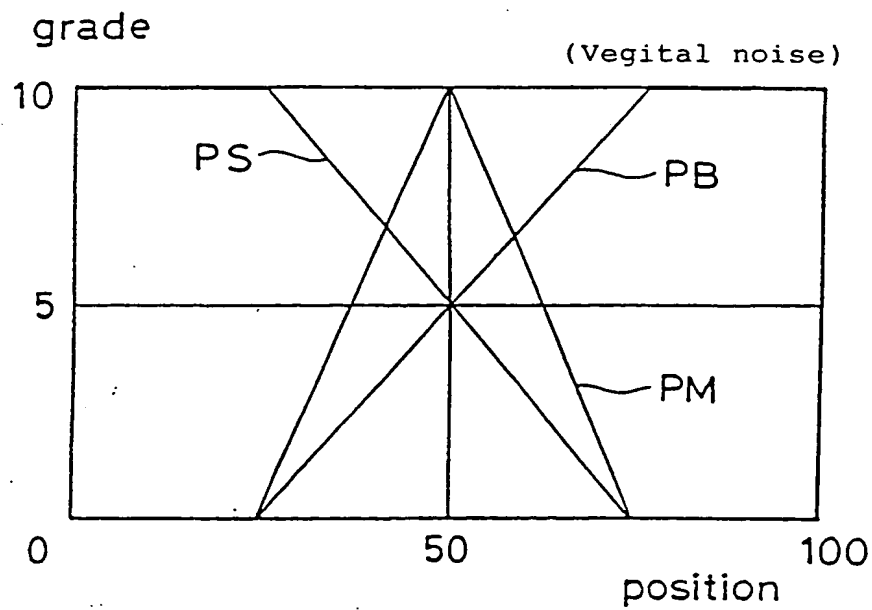


FIG.29

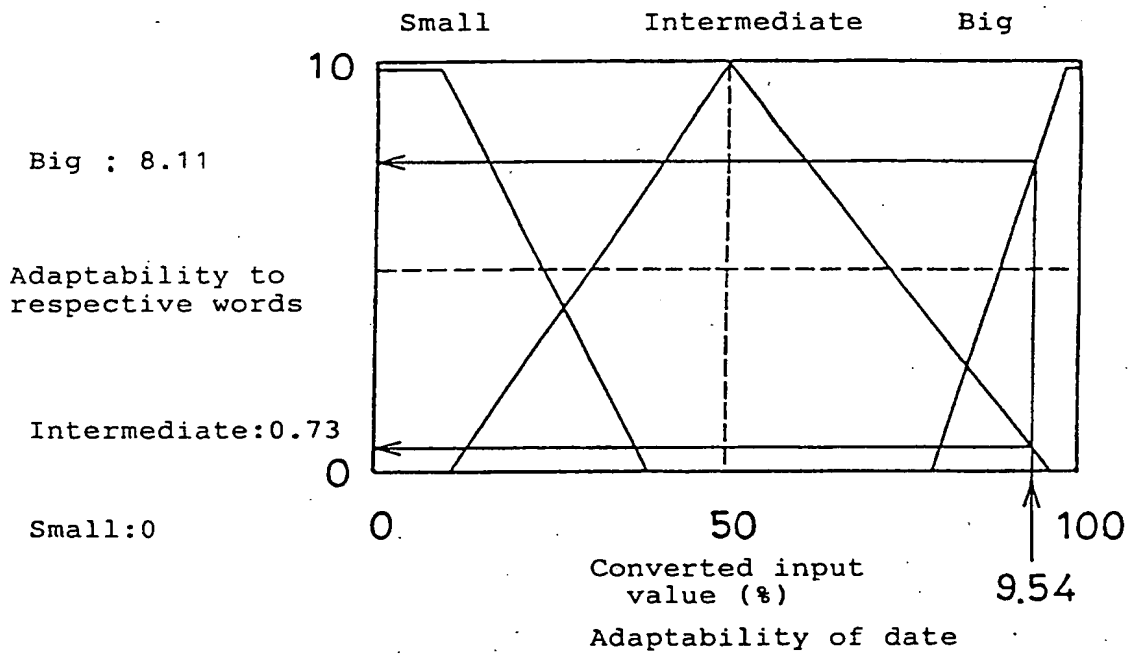


FIG.30

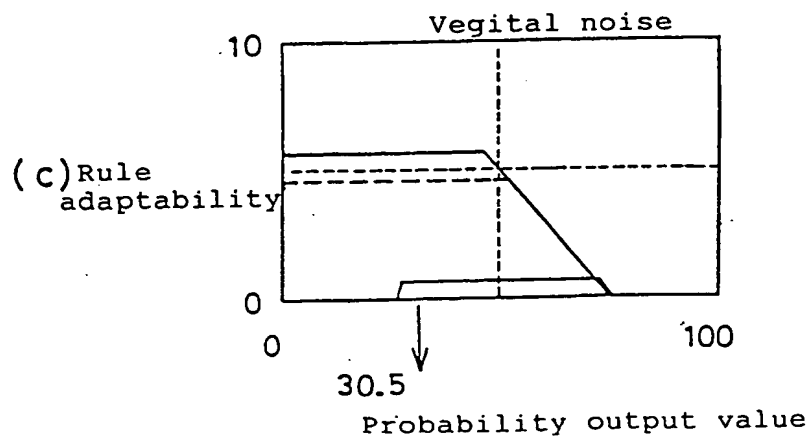
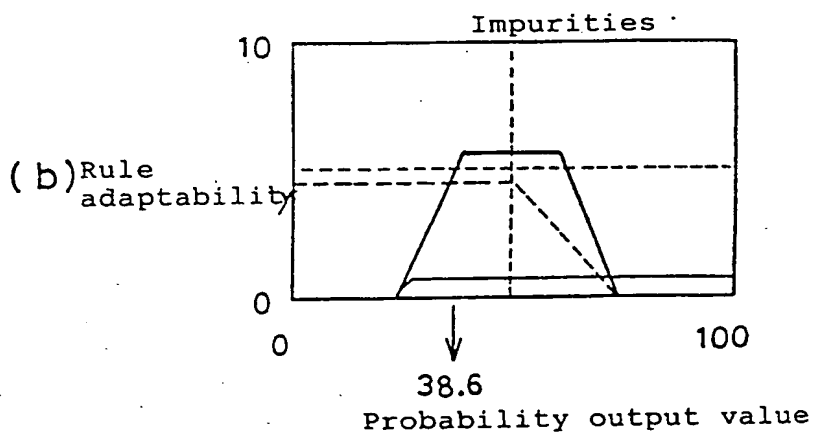
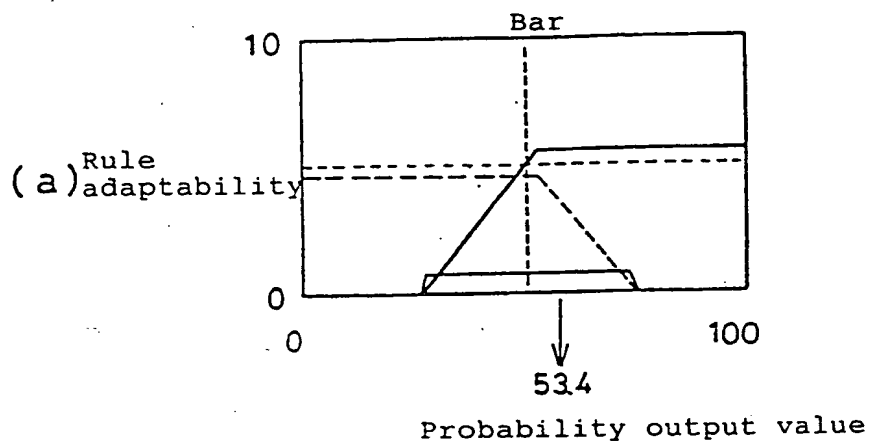


FIG. 31

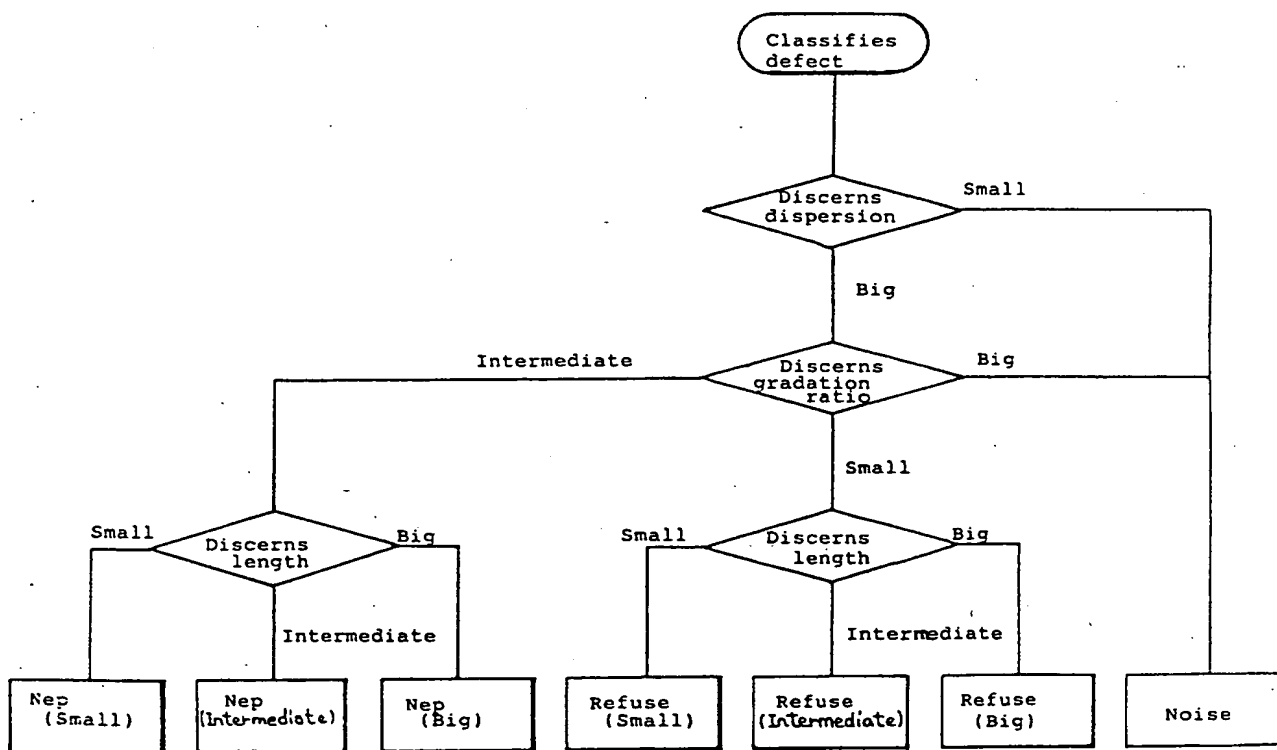


FIG.32

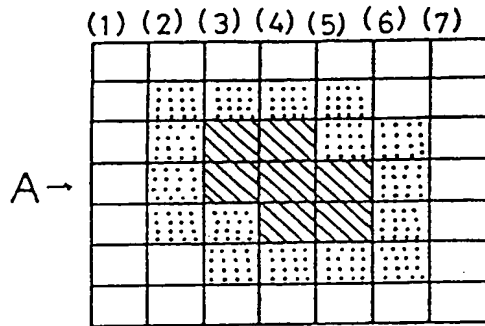


FIG.33

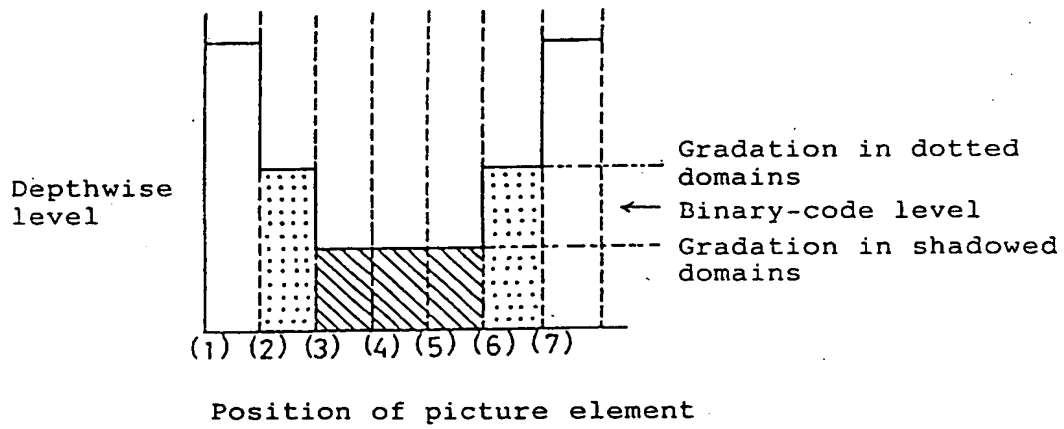
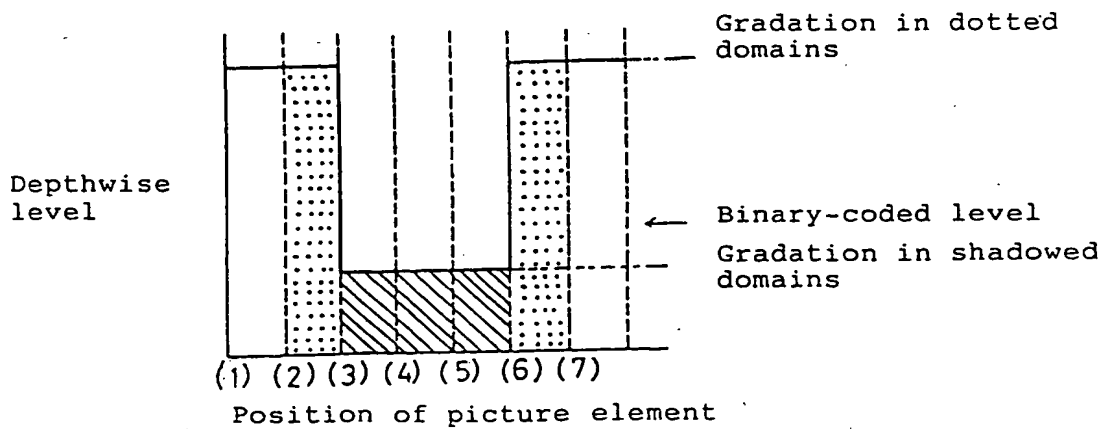


FIG.34



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